# Package 'PCpheno'

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Type Package

Title Phenotypes and cellular organizational units

<b>Version</b> 1.44.0						
Author Nolwenn Le Meur and Robert Gentleman						
<b>Description</b> Tools to integrate, annotate, and link phenotypes to cellular organizational units such as protein complexes and pathways.						
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Imports AnnotationDbi, Biobase, Category, GO.db, graph, graphics, GSEABase, KEGG.db, methods, ScISI, stats, stats4						
Suggests KEGG.db, GO.db, org.Sc.sgd.db						
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# Description

Tools to integrate, annotate and search for associations between phenotypes, protein complexes, and pathways.

# **Details**

Package: PCpheno Type: Package Version: 1.3.1 Date: 2006-03-09

License: The Artistic License, Version 2.0

# Author(s)

N. LeMeur and R. Gentleman

Maintainer: N. LeMeur <nlemeur@fhcrc.org>

#### References

Giaever G, et al. (2002) Functional profiling of the Saccharomyces cerevisiae genome. Nature. 418(6896):387-91. PMID: 12140549

Deutschbauer AM, Jaramillo DF, Proctor M, Kumm J, Hillenmeyer ME, Davis RW, Nislow C, Giaever G. (2005) Mechanisms of haploinsufficiency revealed by genome-wide profiling in yeast. Genetics. 169(4):1915-25.

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Byrne KP, Wolfe KH. (2005) The Yeast Gene Order Browser: combining curated homology and syntenic context reveals gene fate in polyploid species. Genome Res. 15(10):1456-61. PMID: 16169922

#### See Also

ScISI,SLGI

buildFDMat

Build fitness defect contingency matrix

# Description

Function to build a fitness defect contingency matrix where rows correspond to tested genes and columns to experimental conditions.

#### Usage

buildFDMat(data,genenames,condition)

### **Arguments**

data List of 'significant' fitness defect scores and the associated genes at different

experimental conditions.

condition Character vector of the different experimental conditions tested

genenames Character vector of all the tested genes for fitness defect.

### Value

Contingency matrix of genes that present significant fitness defect in different experimental conditions.

### Author(s)

N. LeMeur

# **Examples**

data(GiaeverPheno)
data(GiaeverExpCdt)

data(GiaeverGene)

 $fitness Data <- \ getFDgene (GiaeverPheno, condition=GiaeverExpCdt, cutoff=c(20,100,100), mode="generation", subset GiaeverPhenoM <- \ buildFDMat(data=fitnessData, genenames=GiaeverGene, condition=GiaeverExpCdt[,3])$ 

categoryToEntrezBuilder

Return a list mapping multi-protein complexes IDs to YEAST ids

### **Description**

Return a list mapping multi-protein complexes (category) IDs to the YEAST ids annotated at the category id.

# Usage

```
## S4 method for signature 'CoHyperGParams'
categoryToEntrezBuilder(p)
```

### **Arguments**

р

A subclass of HyperGParams-class

#### **Details**

End users **should not** call this directly. This method gets called from hyperGTest. To add support for a new category, a new method for this generic must be defined. Its signature should match a subclass of HyperGParams-class appropriate for the new category.

#### Value

A list mapping category IDs to YEAST identifiers.

# Author(s)

S. Falcon and N. LeMeur

# See Also

hyperGTest CoHyperGParams-class

CoHyperGParams-class Class "CoHyperGParams"

#### **Description**

A parameter class for representing all parameters needed for running the hyperGTest method with multiprotein complexes.

#### **Objects from the Class**

Objects can be created by calls of the form new("CoHyperGParams", ...).

#### **Slots**

- geneIds: Object of class "ANY": A vector of gene identifiers. Numeric and character vectors are probably the only things that make sense. These are the gene ids for the selected gene set.
- universeGeneIds: Object of class "ANY": A vector of gene ids in the same format as geneIds defining a subset of the gene ids on the chip that will be used as the universe for the hypergeometric calculation. If this is NULL or has length zero, then all gene ids on the chip will be used.
- annotation: A string giving the name of the annotation data package for the chip used to generate the data.
- cateogrySubsetIds: Object of class "ANY": If the test method supports it, can be used to specify a subset of category ids to include in the test instead of all possible category ids.
- categoryName: A string describing the category. Usually set automatically by subclasses. For example "ScISI".
- pvalueCutoff: The p-value to use as a cutoff for significance for testing methods that require it. This value will also be passed on to the result instance and used for display and counting of significant results. The default is 0.01.
- testDirection: A string indicating whether the test should be for overrepresentation ("over") or underrepresentation ("under").

#### Methods

- **hyperGTest** signature(p = "HyperGParams"): Perform hypergeometric tests to assess over-representation of category ids in the gene set. See the documentation for the generic function for details. This method must be called with a proper subclass of HyperGParams.
- geneIds(r), geneIds(r) <- value Accessors for the gene identifiers that will be used as the selected gene list.

#### codeannotation(object) Accessor for annotation

ontology(r) Accessor for GO ontology.

- pvalueCutoff(r), pvalueCutoff(r) <- value Accessor for the p-value cutoff. When setting, value should be a numeric value between zero and one.
- testDirection Accessor for the test direction. When setting, value must be either "over" or "under".
- universeGeneIds(r) accessor for vector of gene identifiers.
- isConditional(r) Returns TRUE if the instance has its conditional flag set

#### Author(s)

S. Falcon and N. LeMeur

#### See Also

HyperGResult-class CoHyperGResult-class hyperGTest

CoHyperGResult-class Class "CoHyperGResult"

### **Description**

This class represents the results of a test for over-representation of genes in a selected gene set based among protein complexes upon the Hypergeometric distribution.

### **Objects from the Class**

Objects is created by calls to the funtion hyperGTest.

#### **Slots**

```
pvalues: "numeric" vector: the ordered p-values for each category term tested.
```

oddsRatios: Object of class "numeric" Odds ratio for each category term tested

expectedCounts: Object of class "numeric" The expected number of genes for each gene term tested

geneCounts: "integer" vector: for each category term tested, the number of genes from the gene set that are annotated at the term.

universeCounts: "integer" vector: for each category term tested, the number of genes from the gene universe that are annotated at the term.

catToGeneId: Object of class "list". The names of the list are category IDs. Each element is a vector of gene IDs annotated at the given category ID and in the specified gene universe.

#### **Extends**

Class "HyperGResultBase", directly.

### Methods

```
geneCounts signature(r = "CoHyperGResult"): return an "numeric" vector: for each category term tested, the number of genes from the gene set that are annotated at the term.
```

```
pvalues signature(r = "HyperGResult"): return a "numeric" vector: the ordered
p-values for each category term tested.
```

```
oddsRatios signature(r = "HyperGResult"): return a "numeric" vector: the odds
  ratio for each category term tested.
```

**expectedCounts** signature(r = "HyperGResult"): return a "numeric" vector: the expected number of genes for each GO term tested.

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universeCounts signature(r = "HyperGResult"): return an "numeric" vector: for each category term tested, the number of genes from the gene universe that are annotated at the term.

**geneIdUniverse** signature(r = "CoHyperGResult"): return a list named by the protein Complexes. Each element of the list is a vector of gene identifiers (from the gene universe) annotated at the corresponding protein complex.

summary signature(r = "CoHyperGResult"): Returns a data.frame summarizing the test result. Optional arguments pvalue and categorySize allow specification of minimum p-value and cateogyrSize, respectively. Optional argument htmlLinks is a logical value indicating whether to add HTML links (useful in conjunction with xtables print method with type set to "html").

# Author(s)

S. Falcon and N. LeMeur

#### See Also

HyperGResultBase-class

### **Examples**

```
data(DudleyPhenoM)
data(ScISIC)
## Select genes sensitive to paraquat
DudleyPhenoL <- apply(DudleyPhenoM,2,function(x) names(which(x==1)))</pre>
paraquat <- DudleyPhenoL[["Paraq"]]</pre>
## Apply a hypergeometric test
params <- new("CoHyperGParams",</pre>
               geneIds=paraquat,
               universeGeneIds=rownames(ScISIC),
               annotation="org.Sc.sgd.db",
               categoryName="ScISIC",
               pvalueCutoff=0.01,
               testDirection="over")
paraquat.complex <- hyperGTest(params)</pre>
## access the p-values
pvalues(paraquat.complex)[1:5]
## Display a summary of the results
summary(paraquat.complex)[,1:4]
```

complexStatus

Complex Status

#### **Description**

Categorize the complex whether or not a complex is composed of a significant number of genes involved in a particular phenotype than expected by chance.

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### Usage

```
complexStatus(data, phenotype, interactome, threshold=0.05)
```

### **Arguments**

data Output from CoHyperG test

phenotype List of gene names inducing an observed phenotype, e.g., list of essential gene

names (see package SLGI)

interactome A binary matrix composed of genes (rows) and biological complexes (columns)

(see package ScISI)

threshold pvalue threshold (default 0.05)

#### **Details**

We form four distinct categories from A to D to characterize how a complex might be involved in a particular phenotype (according to the number of genes it contains and that are involved in a particular phenotype - see also hyperGTest function)

### Value

The returned value is a list with components:

A	"interesting" complexes, complexes with a significant number of interesting genes, i.e., genes that participate to a particular phenotype (at a given p-values threshold)
В	complexes with a NON significant number of interesting genes BUT that SHARE genes with complexes from the A status
С	complexes with a NON significant number of interesting genes AND that DON'T SHARE interesting genes with complexes from cat A

complexes WITHOUT interesting genes, i.e. the one involved in the studied

phenotype

### Author(s)

D

N. LeMeur

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### **Description**

Function to calculate the ratio of genes that characterize a phenotype (observed) among the genes that characterize a biological complex versus the ratio of a set of randomly sampled genes (expected) among the genes that characterize a biological complex.

### Usage

densityEstimate(genename,interactome,perm)

#### **Arguments**

genename Character vector of the gene names that characterize a specific phenotype.

interactome Contingency matrix of genes (rows) and biological complexes (columns) (see

package ScISI)

perm Numeric vector indicating the number of simulations to run to compute the ex-

pected ratios.

#### Value

List of observed and simulated ratios.

### Author(s)

N. LeMeur

# **Examples**

```
data(ScISI)
data(essglist)
essential <- names(essglist)
ScISI <- as.matrix(ScISI)
ratio<- densityEstimate(genename=essential,interactome=ScISI,perm=50)</pre>
```

deResult-class

A class for representing the result of a densityEstimate test.

#### **Description**

A class for representing the result of a densityEstimate test.

### **Slots**

 ${\tt Size:} \ \ Object \ of \ class \ "numeric" \ representing \ the \ size \ of \ the \ cellular \ organizational \ unit \ tested$ 

Observed: Return a "numeric" vector: the observed number of interactions between genes inducing a specific phenotype and each cellular organizational units

Expected: Return a matrix: the expected number of interactions between genes inducing a specific phenotype and each cellular organizational units

DudleyPheno DudleyPheno

#### **Extends**

```
Class "testResult", directly.
```

#### Methods

plot Graphical representation of the test result

### Author(s)

N. LeMeur

### See Also

```
testResult,gtResult,densityEstimate,plot
```

### **Examples**

```
## apply a densityEstimate test
data( DudleyPhenoM)
data(ScISIC)

DudleyPhenoL <- apply(DudleyPhenoM,2,function(x) names(which(x==1)))
pH3 <- DudleyPhenoL[["pH3"]]

perm <- 20
pH3Density <- densityEstimate(genename=pH3, interactome=ScISIC, perm=perm)

## access results
pH3Density@Observed[1:5]

## use of the plot method
plot(pH3Density)</pre>
```

DudleyPheno

List of fitness defect score generated from Dudley et al 2005

### Description

Dudley et al (2005) create a collection of gene-deletion mutants to determine genes that contribute to a particular phenotype in specific environmental conditions. This list is generated from a fitness analysis under 21 different experimental conditions.

# Usage

```
data(DudleyGenelist)
data(DudleyPhenoFull)
data(DudleyPhenoM)
data(DudleySign)
```

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#### **Format**

DudleyGenelist is a character vector of length 814 genes. DudleyPhenoFull is a dataframe of 814 genes by 23 elements. The column contains the yeast gene name. The 22 remaining columns are the experimental conditions (see details). The values obtained in the different condition are the fitness defect scores for the 814 genes sensitive to the experimental condition, as defined by Dudley et al (2005). DudleyPhenoM is a 814 by 22 incidence matrix with rownames corresponding to the genes names and columns to an experimental condition. This matrix contains a 1 in the (i,j) position if the i-th gene is sensitive to the experimental condition of the j-th column; it contains a 0 otherwise.

Yname Yeast systematic gene name

benomyl 15ug/ml benomyl, microtubule function

CaCl2 0.7M calcium chloride, divalent cation

CAD 55uM Cadmium, heavy metal

Caff 2mg/ml Caffeine

cyclohex 0.18ug/ml cycloheximide, protein synthesis

DTT unknown

EtOH YPD + 6% Ethanol

FeLim iron limited, nutrient limited condition

HU 11.4mg/ml Hydroxyurea, DNA replication and repair

HygroB 50ug/ml hygromycin B, aminoglycosides

lowPO4 Low phosphate, nutrient limited condition

MPA 20ug/ml mycophenolic acid, transcriptional elongation

NaCl 1.2M sodium chloride, general stress condition

Paraq 1mM paraquat, oxidative stress

pH3 Low pH, general stress condition

rap 0.1ug/ml rapamycin, protein synthesis

Sorb 1.2M sorbitol, general stress condition

UV 100J/m2 ultra-violet, DNA replication and repair

YPGal 2% galactose, carbon source

YPGly 3% glycerol, carbon source

YPLac 2% lactate, carbon source

YPRaff 2% raffinose, carbon source

DudleySign is a list of dataframe that summaryzes in which complexes the gene related to the phenotype are found, the size of the complexes and the associated p-value. This is the result of applying a Hypergeometric test (see CoHyperGParams-class for more details) and the complexStatus function.

Dudleyresult is a data.frame that summaryzes the number of sensitive genes per condition, how many of those genes are present in the ScISI interactome and the associated p-value. This is the result of applying a Hypergeometric test (see CoHyperGParams-class for more details) and the complexStatus function.

#### Author(s)

N. LeMeur

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#### **Source**

Dudley et al (2005), supplementary information: http://arep.med.harvard.edu/pheno/default.htm

#### References

Aim\'ee Marie Dudley, Daniel Maarten Janse, Amos Tanay, Ron Shamir and George McDonald Church. (2005). A global view of pleiotropy and phenotypically derived gene function in yeast. *Molecular Systems Biology* doi:10.1038/msb4100004

# **Examples**

```
data(DudleyPhenoFull)
data(DudleyPhenoM)
```

getDescr

Get formatted annotation data

# Description

Function to retrieve the annotation of multi-protein complexes or pathways via GO, MIPS or KEGG.

### Usage

```
getDescr(x, database="GO.db")
```

### **Arguments**

x Vector of multi-protein complexes or pathways IDs to be described

database Source of annotation. The database currently available are MIPS, GO.db and

KEGG.db

# Author(s)

N. LeMeur

```
xx <- getDescr(c("MIPS-220","MIPS-260.20","04111"),c("MIPS","KEGG.db"))</pre>
```

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getFDgene	Get fitness defect genes	

#### **Description**

Function to select genes that present a significant growth defect according to the condition(media) or generation time.

### Usage

```
getFDgene(data,condition,cutoff,mode="generation",subset)
```

### **Arguments**

data	List of fitness defect scores for genes tested at different experimental conditions.
condition	Dataframe of experimental conditions
cutoff	Numerical vector of length one or more, defining the threshold of 'significance' for the fitness defect score
mode	Character string defining the base of the selection either 'condition' (media) or 'generation' time, Default=generation.
subset	Numerical vector or list to which apply the different cutoffs.

### Value

Reduced list of gene fitness scores per experimental condition according to the experimental condition or the generation time.

# Author(s)

N. LeMeur

#### References

```
Giaever G. et al. (2002) Functional profiling of the Saccharomyces cerevisiae genome. Nature. 418(6896):387-91. PMID: 12140549
```

```
data(GiaeverPheno)
data(GiaeverExpCdt)
##Select all the genes, in the different experimental conditions, that present a fitness score above 20, 100
fitnessGen <- getFDgene(GiaeverPheno,condition=GiaeverExpCdt,cutoff=c(20,100,100),mode="generation",subset
##Select all the genes, that present a fitness score above 15 and 100
##in the condition set A and B respectively, independently of the generation time
fitnessCondt <- getFDgene(GiaeverPheno,condition=GiaeverExpCdt,cutoff=c(100,15),mode="condition",subset=lis")</pre>
```

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GiaeverPheno

List of fitness defect score generated from Giaever et al 2002

#### **Description**

Giaever et al (2002) create a collection of gene-deletion mutants to determine genes that contribute to a particular phenotype in specific environmental conditions. This list is generated from a fitness analysis under six different experimental conditions.

### Usage

```
data(GiaeverPheno)
data(GiaeverGene)
data(GiaeverExpCdt)
```

#### **Format**

GiaeverPheno is a list with 31 elements. The name of each element is a experimental condition (see details). The value of each element are the fitness defect scores for the genes sensitive to the experimental condition, as defined by Giaever et al (2002).

GiaeverGene Vector of the systematic gene names of the 5898 tested genes. Note that some updates have been made for the list to be consistent with *Saccharomyces Genome Database*.

GiaeverExpCdt is a 3 columns dataframe with fileID from which the data were extracted, the generation time (growth time) and the condition (media).

```
gen. generations
rep. replicate
ypg5a,ypg5b yeast/peptone/galactose 5 gen. rep. a and b ==> carbone source
ypg15a ypg15b yeast/peptone/galactose 15 gen. rep. a and b ==> carbone source
sorbitol5a sorbitol5b 1.5M Sorbitol 5 gen. rep. a and b ==> sugar, osmotic stress
sorbitol20a sorbitol15b 1.5M Sorbitol 20 and 15 gen. rep. a and b respectively==> sugar, osmotic
NaCl5a NaCl5b 1M NaCl 5 gen. rep. a and b ==> salt, osmotic stress
NaCl15a NaCl15b 1M NaCl 15 gen. rep. a and b ==> salt, osmotic stress
lysM5a lysM5b lysine minus 5 gen. rep. a and b ==> lack of required AA
thM5a threonine minus 5 gen. rep. a ==> lack of required AA
trpM5a trpM5b tritophanee minus 5 gen. rep. a ==> lack of required AA
minimalPlus5a minimalPlus5b minimal + histidine/leuvine/uracile 5 gen. rep. a and b
minimalPlus15a minimalPlus15b minimal + histidine/leuvine/uracile 15 gen. rep. a and b
minimalC5a minimalC5b minimal complete 5 gen. rep. a and b
nystatin5a nystatin5b Nystatin 5 gen. rep. a and b ==> antifungal drug
nystatin15a nystatin15b Nystatin 5 gen. rep. a and b ==> antifungal drug
pH8g5a pH8g5b pH 8 5 gen. rep. a and b ==> alkali stress
pH8g15a pH8g20b pH 8 15 and 20 gen. rep. a and b respectively ==> alkali stress
```

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Note: in their study they confound the 15 and 20 generations.

Giaeverresult is a data frame that summaryzes the number of sensitive genes per condition, how many of those genes are present in the ScISI interactome and the associated p-value. This is the result of applying a Hypergeometric test (see CoHyperGParams-class for more details) and the complexStatus function.

#### Author(s)

N. LeMeur

#### Source

Giaever et al (2002), supplementary information: http://genomics.lbl.gov/YeastFitnessData/websitefiles/cel\_index.html Saccharomyces Genome Database (last update 03/17/06): http://www.yeastgenome.org/

#### References

Giaever G et al (2002) Functional profiling of the Saccharomyces cerevisiae genome. Nature. 418(6896):387-91. PMID: 12140549

#### **Examples**

```
data(GiaeverPheno)
data(GiaeverExpCdt)
data(GiaeverGene)
```

graphTheory

Graph theory to test associations between two or more relationships

### Description

Graph theory approach associated with a permutation test to evaluate whether the number of associations is unexpectedly large.

### Usage

```
graphTheory(genename, interactome, perm)
```

### **Arguments**

genename A vector a gene names that are associated with a particular phenotype

interactome A binary matrix composed of genes (rows) and biological complexes (columns)

(see package ScISI)

perm Numeric, number of permutation run

#### **Details**

We form two distinct graphs where the set of nodes are the proteins (genes) in the organism. In one graph we create edges between genes if the two genes are members of one, or more, protein complexes. In the second graph we create an edge between all genes that are associated to a particular phenotype. We then construct a third graph on the same node set, but where there is an edge in this graph only if there is an edge in both of the first to graphs. We count the number of edges in the third and test by permutation whether the number of edges is unexpectedly large.

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#### Value

The returned value is a list with components:

edgeCount Number of associations observed between the genes that are linked to a particu-

lar phenotype and the given interactome.

edgeSimul Number of associations if the genes that are linked to a particular phenotype are

randomly distributed across the given interactome

p.value Returned p.value

#### Author(s)

R. Gentleman and N. LeMeur

#### References

Balasubramanian, R., LaFramboise, T., Scholtens, D., Gentleman, R. (2004) A graph-theoretic approach to testing associations between disparate sources of functional genomics data. Bioinformatics, 20(18), 3353-3362.

#### **Examples**

```
data(ScISI)
data(essglist)
ans <- graphTheory(names(essglist), ScISI, perm=3)</pre>
```

gtResult-class

A class for representing the result of a graphTheory test.

### **Description**

A class for representing the result of a graphTheory test.

### **Slots**

Pvalue: Object of class "numeric"

Observed: Return a "numeric" vector: the observed number of interactions between genes inducing a specific phenotype and each cellular organizational units

Expected: Return a matrix: the expected number of interactions between genes inducing a specific phenotype and each cellular organizational units

### **Extends**

```
Class "testResult", directly.
```

#### Methods

plot Graphical representation of the test result

#### Author(s)

N. LeMeur

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#### See Also

```
testResult,deResult,plot
```

#### **Examples**

```
## apply a densityEstimate test
data(DudleyPhenoM)
data(ScISIC)

DudleyPhenoL <- apply(DudleyPhenoM,2,function(x) names(which(x==1)))
NaCl <- DudleyPhenoL[["NaCl"]]

perm <- 20
NaClGraph <- graphTheory(genename=NaCl, interactome=ScISIC, perm=perm)

## access results
slotNames(NaClGraph)
NaClGraph@Pvalue[1:5]

## use of the plot method
plot(NaClGraph)</pre>
```

ΗI

Data from Deutshbauer et al. (2005)

#### **Description**

Mechanisms of Haploinsufficiency revealed by Genome-Wide Profiling in Yeast (Deutshbauer et al.,2005)

### Usage

```
data(HI)
```

### **Details**

HI stands for haploinsufficient. The dataframe is composed of:

```
orf Yeast ORF, systematic namegene Yeast common gene name of the corresponding ORFgo GO terms
```

### **Source**

```
http://www.sciencemag.org/cgi/data/303/5659/808/DC1/1
```

# References

Deutschbauer AM, Jaramillo DF, Proctor M, Kumm J, Hillenmeyer ME, Davis RW, Nislow C, Giaever G. (2005) Mechanisms of haploinsufficiency revealed by genome-wide profiling in yeast. Genetics. 169(4):1915-25.

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#### **Examples**

data(HI)

KastenmayerRaw

Data from Kastenmayer et al. 2006

### **Description**

Kastenmayer et al. (2006) undertook the first functional studies of small open reading frames (sORFs) in any system, using the model eukaryote Saccharomyces cerevisiae. Phenotypic analyses of the new gene-deletion strains identified 22 sORFs required for haploid growth, growth at high temperature, growth in the presence of a non-fermentable carbon source, or growth in the presence of DNA damage and replication-arrest agents.

### Usage

data(KastenmayerRaw)

#### **Format**

Kastenmayer is a 5 columns dataframe.

**SYSTEMATIC** Systematic name of the sORF.

**COMMUN** Commun name of the sORF.

Length Length of the small ORF sequence in number of amino acids.

Evidence Experimental source of the data

refHomology Bibliographical evidence of reported homology.

**Kocollection** Bibliographical evidence of reported homology.

**ESSENTIAL** Indicates if the sORF knockout is essential. A blank in this column indicates that the knockout is not-essential, if available.

**GFPTAP** "GFP" or "TAP" signifies that sORF was detected by the indicated technique. "both" indicates that sORF was detected both as a TAP-tagged and GFP-tagged protein. "None" indicates that sORF was not detected by either method. Empty field indicates that sORF was not tested

**UPTAG** Sequence of the upstream primer.

**DOWNTAG** Sequence of the downstream primer.

#### Author(s)

N. LeMeur

### **Source**

Kastenmayer et al. (2006), supplementary information: http://www.genomeresearch.org or ftp://genome-ftp.stanford.edu/pub/yeast/systematic\_results/phenotypes

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#### References

Kastenmayer JP, Ni L, Chu A, Kitchen LE, Au WC, Yang H, Carter CD, Wheeler D, Davis RW, Boeke JD, Snyder MA, Basrai MA. (2006) Functional genomics of genes with small open reading frames (sORFs) in S. cerevisiae. Genome Res. 16(3):365-73. PMID: 16510898

### **Examples**

```
data(KastenmayerRaw)
str(KastenmayerRaw)
```

KEGG2SCISI

Mapping between KEGG and ScISI

# Description

Count the number of genes shared between a KEGG pathway and a protein complex from the ScISI interactome.

### Usage

```
KEGG2SCISI(pw, pc, pcMat, pwMat)
```

### **Arguments**

pw list of pathway names
pc list of complex names
pwMat pathway incidence matrix
pcMat complex incidence matrix

### Value

matrix

#### Author(s)

N. LeMeur

### See Also

ScISI KEGG

```
data(ScISIC)
## Mapping from Yeast genes to KEGG pathways.
KeggMat <- PWAmat("org.Sc.sgd")
KEGG2SCISI(pw = colnames(KeggMat)[1:5], pc = colnames(ScISIC)[1:5], pwMat =
KeggMat, pcMat =ScISIC)</pre>
```

20 LesageRaw

LesageRaw

Data from Lesage et al. 2005

#### **Description**

Lesage et al. (2005) assembled a network of 316 interactions among 163 genes using deletion mutants in CHS1, CHS3, CHS4, CHS5, CHS6, CHS7 and BNI4 in a synthetic genetic array analysis.

#### Usage

data(LesageRaw)

#### **Format**

LesageRaw is a 5 column dataframe.

**SYSTEMATIC** Systematic gene names. NOTE: All mutants are isogenic to BY4741 (MATa his3\u0394 leu2\u0394 met15\u0394 ura3\u0394) except anp1\u0394 and mnn9\u0394 that are isogenic to BY4742 (MAT\u03b1 his3\u0394 leu2\u0394 lys2\u0394 ura3\u0394).

**COMMUN** Commun gene names.

**CFW** Mutants showing increased, decreased or wild type sensitivity to Calcofluor white are scored s, r, or wt, respectively.

**ChitinLevel** Chitin level (nmole GlcNAc/mg dry weight). Values are an average of at least three independent determinations. Values statistically higher and lower than wild type (p < 0.01) are highlighted in red and green, respectively.

**ChitinLevel.SD** Standard deviation of the average of at least three independent determinations of Chitin level.

#### Author(s)

N. LeMeur

#### **Source**

Lesage et al. (2005), supplementary information: http://www.biomedcentral.com/1471-2156/6/8/suppl/S2 or ftp://genome-ftp.stanford.edu/pub/yeast/systematic\_results/phenotypes

#### References

Lesage G, Shapiro J, Specht CA, Sdicu AM, Menard P, Hussein S, Tong AH, Boone C, Bussey H. (2005) An interactional network of genes involved in chitin synthesis in Saccharomyces cerevisiae.BMC Genet.6(1):8. PMID: 15715908

### **Examples**

data(LesageRaw)
str(LesageRaw)

OsterbergRaw 21

OsterbergRaw

Data from Osterberg at al. 2006

#### **Description**

Osterberg at al. (2006) report growth phenotypes in yeast for a strain collection over-expression ~600 C-terminal tagged integral membrane proteins growth both under normal and three different stress conditions.

#### Usage

data(OsterbergRaw)

#### **Format**

OsterbergRaw is a 17 columns dataframe.

SYSTEMATIC Systematic gene names of the studied membrane protein

**COMMUN** Commun gene names of the studied membrane protein

- **TMHMM.C** The topology predicted by TMHMM (TransMembrane prediction using Hidden Markov Models) using the experimentally assigned C-terminal location for the protein as a constraint. The topology is represented in the format Location of N-terminus TMhelices Location of C-terminus (i and o stand for in and out respectively)
- **WesternBlot** Protein expression levels (arbitrary units), estimated from the band intensity and normalized to the internal standard on each Western blot.
- **Bands** Proteins detected as two distinct bands with different molecular mass on the Western blot analysis. category 1 indicates that both bands were insensitive to Endo H digestion, 2 indicates the higher molecular mass was shifted down upon Endo H digestion on SDS/PAGE and one band was predominant compared to the other, and 3 indicates that a higher molecular mass band shifted down upon Endo H digestion on SDS/PAGE and both bands were equal intensity on Western blot
- **Toxicity** Toxicity index from Spoko et al. (2006). The index varies between 1 and 5, where 1 means the strain is dead, and 5 indicates no difference in growth rate compared with the wild type strain.
- **sign.norm** Over-expression strains that show a significant (P<0.001) growth rate phenotype (LSCrate) in synthetic defined medium conditions (Warringer et al., 2003). An average of two replicates is given. Strains that do not show a significant difference in doubling time compared with the wild-type strain are indicated by 0.
- **all.norm** Phenotypes (significant or not) of over-expression strains (LSCrate) in synthetic defined medium conditions (Warringer et al., 2003). An average of two replicates is given.
- **sign.NaCl** Over-expression strains that show a significant (P<0.001) growth rate phenotype (LPI-rate) in NaCl. An average of two replicates is given. Strains that do not show a significant difference in doubling time compared with the wild-type strain under NACL stress are indicated by 0.
- **all.NaCl.LSC** Phenotypes (significant or not) of over-expression strains (LSCrate) in NaCl (Warringer et al., 2003). An average of two replicates is given.
- **all.NaCl.LPI** Phenotypes (significant or not) of over-expression strains (LPIrate) in NaCl (Warringer et al., 2003). An average of two replicates is given.

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**sign.caff** Over-expression strains that show a significant (P<0.001) growth rate phenotype (LPI-rate) in caffeine. An average of two replicates is given. Strains that do not show a significant difference in doubling time compared with the wild-type strain under caffeine stress are indicated by 0.

- **all.caff.LSC** Phenotypes (significant or not) of over-expression strains (LSCrate) in caffeine (Warringer et al., 2003). An average of two replicates is given
- **all.caff.LPI** Phenotypes (significant or not) of over-expression strains (LPIrate) in caffeine (Warringer et al., 2003). An average of two replicates is given.
- **sign.paraq.** Over-expression strains that show a significant (P<0.001) growth rate phenotype (LPI-rate) in paraquat. An average of two replicates is given. Strains that do not show a significant difference in doubling time compared with the wild-type strain under paraquat stress are indicated by 0.
- **all.paraq.LSC** Phenotypes (significant or not) of overexpression strains (LSCrate) in paraquat (Warringer et al., 2003). An average of two replicates is given.
- **all.paraq.LPI** Phenotypes (significant or not) of overexpression strains (LPIrate) in paraquat (Warringer et al., 2003). An average of two replicates is given.

#### Author(s)

N. LeMeur

#### Source

Osterberg et al (2006), supplementary information: http://www.pnas.org/content/vol0/issue2006/images/data/0604078103/DC1/04078Table1.xls or ftp://genome-ftp.stanford.edu/pub/yeast/systematic\_results/phenotypes

#### References

Osterberg M, Kim H, Warringer J, Melen K, Blomberg A, von Heijne G. (2006) Phenotypic effects of membrane protein overexpression in Saccharomyces cerevisiae. PNAS. 103(30):11148-53. PMID: 16847257

### **Examples**

data(OsterbergRaw)
str(OsterbergRaw)

overlap

Count the number of proteins shared by protein complexes

### **Description**

Count the number of proteins shared by protein complexes

#### Usage

overlap(interactome)

plot 23

### **Arguments**

interactome Binary matrix composed of genes (rows) and biological complexes (columns)

(see package ScISI)

#### Value

The returned value is a data frame with components:

C1 Name of the first biological complex
C2 Name of the second biological complex

nbSharedProt Number of proteins in common

#### Author(s)

N. LeMeur

#### See Also

ScISI

### **Examples**

```
xx = cbind("a"=c(0,1,1,1),"b"=c(1,1,0,1))
overlap(xx)
```

plot

Graphical method to represent the result of the density or graph test.

### **Description**

a plot method for deResult and gtResult objects.

### Usage

```
## S4 method for signature 'deResult' plot(x,...) ## S4 method for signature 'gtResult' plot(x,...)
```

#### **Arguments**

x the deResult or gtResult object to plot.... general commands to be sent to plot.

### **Details**

The plot generated from a deResult object is a set of density plots.

The plot generated from a gtResult object is a histogram.

#### Author(s)

N. LeMeur

24 ppiInteract

#### See Also

ScISI

### **Examples**

```
data(ScISI)
data(essglist)
essential <- names(essglist)
ScISI <- as.matrix(ScISI)
ratio<- densityEstimate(genename=essential,interactome=ScISI,perm=50)
plot(ratio)</pre>
```

ppiInteract

Test the association between AP-MS data and phenotype

### **Description**

Test the association between AP-MS data and phenotype data via a graph and permutation model.

### Usage

```
ppiInteract(genename, expGraph, bait, prey, perm=10)
```

# **Arguments**

genename Genes associated to a phenotype

expGraph A graphNEL object (a direct graph instance of classgraph). The nodes are the

union of viable baits (VB) and viable prey (VP) of the experiment (see package

ScISI)

Proteins which was sampled as a bait in the binary relationship

Proteins which was sampled as a prey in the binary relationship

perm Number of permutation

### Value

The returned value is a list:

Observed Values

Expected Expected values after each permutation

### Author(s)

R. Gentleman and N. LeMeur

### See Also

ScISI

reduceM 25

### **Examples**

reduceM

Resize a matrix

# Description

Resize a matrix to the number of rows commun to a vector.

# Usage

```
reduceM(x, mat, threshold=0)
```

# **Arguments**

x Character or numeric vector.

mat Matrix sharing rownames with the supplied vextor x.

threshold Threshold upon column. Only the columns with a colSums above the threshold

are kept.

### Value

Resized matrix.

### Author(s)

N. LeMeur

```
mat <- matrix(c(1:25), nrow = 5, ncol = 5, dimnames = list(c(LETTERS[1:5]), c(1:5))) xx <- LETTERS[c(2, 4, 5)] reduceM(xx, mat)
```

26 testResult-class

SGDphenoL

Saccharomyces Genome Database list of phenotypic data

### **Description**

Saccharomyces Genome Database list of phenotypes and associated genes from several published experiments (last update 2006).

# Usage

```
data(SGDphenoL)
```

#### **Format**

SGDphenoL is a list of phenotypes. Under each phenotype is listed the genes that potentially induce that phenotype. A binary matrix can be built from that list using the list2Matrix function from the Rintact package.

### Author(s)

N. LeMeur

#### **Source**

SGD, supplementary information: http://www.yeastgenome.org/

### **Examples**

data(SGDphenoL)

testResult-class

A virtual class for representing the result of a test.

### **Description**

The testResult class is the virtual base class for all result objects of the densityEstimate and graphTheory tests proposed in PCpheno.

#### **Objects from the Class**

A virtual Class: No objects may be created from it.

#### **Slots**

Observed: Return a "numeric" vector: the observed number of genes or interactions within each cellular organizational units

Expected: Return a numeric or a matrix: the expected number of genes or interactions within each cellular organizational units

truncName 27

### Methods

No methods defined with class "testResult" in the signature.

# Author(s)

N. LeMeur

# See Also

```
gtResult,deResult
```

truncName

Truncate character strings

# Description

Truncate character strings

# Usage

truncName(x,n)

# Arguments

x Character string

n Maximum length (in characters)

# Value

Character string

# Author(s)

N. LeMeur

```
xx <- "Anticonstitutionnelement is a family name"
truncName(xx,5)</pre>
```

28 YEASTOHNOLOG

YEASTOHNOLOG

List of ohnolog gene pairs from Byrne, K.P and Wolfe, K.H (2005)

#### **Description**

List of 551 paralogous *Saccharomyces cerevisiae* gene pairs formed by Whole Genome Duplication (WGD) or ohnolog pairs.

#### Usage

data(YEASTOHNOLOG)

#### **Format**

YEASTOHNOLOG is a dataframe of 551 paired genes. The first two columns are the ohnolog gene pairs (systematic gene names). The third column is an index (numeric) of the rate of sequence evolution. The last two columns define the chromosome location.

#### **Details**

Gene1 Gene2 Systematic gene names of the ohnolog pairs

**Ka** Coefficient that represents the extent of non-synonymous sequence divergence between each ohnolog pairs (Yang and Nielsen, 2000). The highest is the coefficient the fastest the 2 elements of a pair have diverged.

**ChrG1 ChrG2** Chromosome location of the each element of a pair. Note that repeat of the same chromosome locations shared by a set of pairs define a block of duplication.

#### Author(s)

N. LeMeur

### Source

Byrne, KP and Wolfe KH (2005), Table 2 of supplementary information and Scerevisiae\\_genome.tab file, chromosome location, from YGOB http://wolfe.gen.tcd.ie/ygob/ (last update 03/20/06)

#### References

Byrne KP, Wolfe KH. (2005) The Yeast Gene Order Browser: combining curated homology and syntenic context reveals gene fate in polyploid species. Genome Res. 15(10):1456-61. PMID: 16169922

### **Examples**

data(YEASTOHNOLOG)
str(YEASTOHNOLOG)

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